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Association of the TGFβ gene family with microenvironmental features of gastric cancer and prediction of response to immunotherapy

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In the complex tumor microenvironment, TGFB is a pleiotropic cytokine involved in regulating cellular processes such as cancer cell proliferation, apoptosis and metastasis. TGFB defines three subtypes (TGFB1, TGFB2, and TGF β 3), of which TGF β is highly expressed in many cancers, especially those showing high dissemination potential. In addition, increased expression of TGF_β in multiple cancers is usually positively correlated with epithelial mesenchymal transition (EMT) and coordinated with the expression of genes driving EMT-related genes. TGF β signaling in the tumor microenvironment inhibits the antitumor function of multiple immune cell populations, including T cells and natural killer cells, and the resulting immunosuppression severely limits the efficacy of immune checkpoint inhibitors and other immunotherapeutic approaches. As a major pathway to enhance the efficacy of cancer immunotherapy effects, the role of TGF β signaling inhibitors have been evaluated in many clinical trials. However, the potential functions and mechanisms of TGFB1, TGFB2 and TGFB3 in gastric cancer progression and tumor immunology are unclear. In this study, we comprehensively analyzed TGF_{β1}, TGF_{β2} and TGF_{β3} and gastric cancer microenvironmental features, including immune cell infiltration, EMT, hypoxia, mutation, immunotherapy and drug treatment, based on HMUCH sequencing data (GSE184336) and public databases. We also validated the protein expression levels of TGFB in gastric cancer tissues as well as the role of TGFB factor in cytology experiments. This report reveals the important role of the TGF β gene family in gastric cancer and provides possible relationships and potential mechanisms of TGF β in gastric cancer.

KEYWORDS

 $TGF\beta$, tumor microenvironment, epithelial-mesenchymal transition, hypoxia, immunotherapy, gastric cancer

Introduction

In the past decades, great progress has been made in surgical treatment techniques and adjuvant therapy for gastric cancer, but its prognosis is still not ideal, and cancer recurrence often occurs (1, 2). With the development of tumor biology, more and more studies have shown that the occurrence, development and metastasis of tumors are closely related to the tumor microenvironment (TME). TME is both the cause and the result of tumor development, therefore, understanding the characteristics of TME and their changing features at different stages of tumor development is of great significance for tumor diagnosis and treatment.

TGF β is a powerful cytokine in the tumor microenvironment that regulates most cellular behaviors in the TME. In general, TGFB enhances immune tolerance and suppresses inflammation, mechanisms that are often exploited during tumor evolution to evade surveillance and combat by the immune system (3). In addition, TGFB molecules can also play an important role in promoting EMT, phenotypic transformation of CAFs, angiogenesis and maintaining tumor stemness in tumors (4-7). The close link between TGFB signaling pathway and tumor development makes TGFB signaling pathway a possible new target for tumor therapy. As a new therapeutic strategy, a growing number of drugs aim to block the activation of TGFB signaling, including TGFB isoform-specific blocking antibodies, given the favorable toxicity profile of these TGFB inhibitors, as well as their ability to modulate immune checkpoint activity, $TGF\beta$ Inhibitors can synergistically enhance the efficacy of various immunotherapies (8, 9).

Currently, research on the TGF β molecule family (TGF β 1, TGF β 2 and TGF β 3) has focused on TGF β 1 and TGF β 2 molecules, while studies related to TGF β 3 molecules are still relatively rare (10–13). To comprehensively analyze the potential functions and roles of the TGF β molecular family in gastric cancer, we performed a comprehensive analysis of TGF β in HMUCH sequencing data (GSE184336) and multiple gastric cancer public datasets, and validated it in gastric cancer tissues and gastric cancer cells.

Materials and methods

Sample collection and data collection

We collected frozen tissues from 231 patients who underwent radical gastric cancer surgery from 2016 to 2019 at the Harbin Medical University Cancer Hospital. Inclusion criteria were preoperative CT examination or gastroscopy and pathological examination to confirm gastric cancer, excluding patients with preoperative chemotherapy, radiotherapy and other adjuvant treatments. All gastric cancer tissues were examined independently by two certified pathologists to confirm the histological type. High-throughput sequencing data of the transcriptome of gastric cancer samples were uploaded to the GEO dataset (GSE184336). This study complied with the requirements of the Research Ethics Committee of the Affiliated Cancer Hospital of Harbin Medical University (2019-164-R).

In addition, gene expression data for pan-cancer were downloaded from the public database The Cancer Genome Atlas (TCGA), which for gastric adenocarcinoma (STAD) also includes copy number variants, mutation data (MAF) and corresponding clinicopathological data. In the TNM staging system, T refers to the condition of the primary tumor site and the extent of adjacent tissue involvement, N refers to regional lymph node involvement, and M refers to distant metastasis. Abbreviations for Pan-cancer showed in Supplementary Table 1. The count data used for differential analysis of genes between different groups, and TPM data used to compare the expression levels of different genes. Download the GSE15459, GSE26253, GSE29272, GSE34942, GSE62254, GSE63089 and GSE84437 gastric cancer datasets from the Gene Expression Omnibus (GEO) database for further analysis (14 - 20).

Gene set variation analysis

In the GSE184336 dataset we performed GSVA calculations using the R package "GSVA" (21). GSVA is a non-parametric and unsupervised algorithm commonly used to estimate changes in pathway and biological process activity in samples of expression datasets. The "c2.cp.kegg.v7.1.symbols" and "h.all.v7.1.entrez.gmt" gene sets were downloaded from the MSigDB database (http://www.gsea-msigdb.org/)and used to run GSVA. The optimal cutoff values of TGF β 1, TGF β 2 and TGF β 3 were determined according to the ROC curve and grouped. The limma package was used to analyze enrichment score (ES) matrices between different TGF β subgroups to explore biological differences between patients in different TGF β subgroups.

Gastric cancer microenvironment assessment

To assess the gastric cancer microenvironment, the stromalscore, immunescore and ESTIMATEScore were calculated using the ESTIMATE package, where tumor purity = $\cos (0.6049872018 + 0.0001467884 * ESTIMATEScore)$ (22). The ssGSEA algorithm assessed the EMT scores and the composition of different types of immune cells in patients with gastric cancer (18, 23).

Mutation analysis

Gistic 2.0 software was used to identify somatic copy number changes (24). The R package maftools was used to analyze MAF mutation information (25). The SCNA module in the TIMER database (http://timer.cistrome.org/) compared the relationship between different somatic copy number changes of TGF β and immune cell infiltration (26).

Prediction of immunotherapy and chemotherapy

In this study, we used the submap module in the GenePattern cloud server (https://cloud.genepattern.org/gp) to analyze the response of patients with different TGF β subgroups to immunotherapy (PD-1 and CTLA-4) (27). In addition, we also used the ImmuCellAI database (http://bioinfo.life.hust.edu. cn/ImmuCellAI) to predict the response of immune checkpoint blockade (ICB) therapy (anti-PD1 or anti-CTLA4 therapy) (28). Patient sensitivity to drugs was assessed according to the Genomics of Drug Sensitibity in Cancer (GDSC) database, and the half maximal inhibitory concentration (IC50) was quantified and analyzed by the R package pRRophetic (29).

Western blot analysis and immunohistochemistry

Six pairs of frozen samples of gastric cancer and adjacent normal tissues (more than 5 cm from the tumor margin) were selected, and total protein was extracted with lysis buffer containing protease inhibitors, and the concentration was determined. PVDF membranes (Merck Millipore) were blocked with 5% skim milk powder and incubated overnight at 4°C with primary antibodies (TGF β 1, AF1027; TGF β 2, 19999-1-AP and TGF β 3, 18942-1-AP). Gastric cancer tissue was paraffin-embedded and cut into 5 mm thick sections, and immunohistochemistry (IHC) staining was performed as previously described (30).

Conditional culture of gastric cancer cells

Gastric cell line (HGC-27) was purchased from Shanghai Institutes for Biological Sciences, Chinese Academy of Sciences, and cultured at 37°C in RPMI-1640 medium containing 10% 100 U/ml penicillin and 100 U/ml streptomycin. The active recombinant proteins TGF β 1 (Proteintech, HZ-1011), TGF β 2 (Proteintech, HZ-1092) and TGF β 3 (Proteintech, HZ-1090) were solubilized and added to normal medium to adjust the concentration of TGF β factor to 10ng/mL. The gastric cancer cells cultured with TGF β active protein were added as the TGF β -treated group, and human serum albumin (HSA) of equal quality was added as the control group and incubated in the cell incubator for 48 hours.

Transient transfection

HGC27 cells were inoculated in 6-well plates and when the cell growth density reached 40-50%, transfection was performed using jetPRIME transfection reagent (Polyplus Transfection, France) according to the manufacturer's protocol. siRNA sequences were shown in Supplementary Table 2.

qRT-PCR and transwell assay

Total RNA was extracted from GC cell lines using TRIzol (Transgen Biotechnology, China). RNA was reverse-transcribed into cDNA using reverse transcription system (Promega, USA). The SYBR-Green (Vazyme) mixed system was then assayed in a LightCycler[®] 480 Real-Time PCR System (Roche) to analyze the cDNA expression levels. The 2- $\Delta\Delta$ Ct method was used to calculate the relative expression levels (31). Transwell was applied based on the previous method (32). The PCR primers were shown in Supplementary Table S3.

Statistical analysis

The chi-square test was used to analyze the association between different TGF β subgroups and clinicopathological parameters. Kaplan-Meier (KM) survival curves were used to compare the survival analysis of different subgroups, and univariate cox regression analysis was also performed on TGF β . Pearson correlation analysis was performed to test the relationship between the different variables. The results of the experimental data were expressed as mean ± SD (standard deviation). Statistical analyses were performed in R software (https://www.r-project.org/, version 4.01) and Graphpad Pris 7.0. The P-value < 0.05 was considered statistically significant.

Results

Gene characteristics and expression levels of $\mathsf{TGF}\beta$

We first showed the chromosomal location of the TGF β gene family and related genes (GeneMANIA database, https://genemania.org/), and compared the copy number changes (gain, none and loss) of different genes (Figures 1A, B). In gastric



cancer dataset STAD, GSE63089 and dataset GSE184336, the expression levels of TGF β 1, TGF β 2 and TGF β 3 were higher in cancer tissues than in normal tissues adjacent to the cancer (Figures 1C–E).

Survival analysis of $\text{TGF}\beta$ in pan-cancer and gastric cancer

The results of the Cox univariate analysis in the TCGA database showed that TGF β 1, TGF β 2 and TGF β 3 were prognostic risk factors in most cancers (BLCA, PAAD, KIRP, LIHC, MESO, COAD, THCA, HNSC, LGG, ACC and UVM), notably TGF β 1, TGF β 2 and TGF β 3 were all poor prognostic factors in STAD. The KM survival analysis results showed that patients with high expression levels of TGF β 1, TGF β 2 and TGF β 3 (GBM, LUSC, TGCT, SARC, BLCA, PAAD, KIRP, LIHC, MESO, COAD, THCA, HNSC, READ, LGG, ACC and UVM) had a shorter survival (Figure 1F).

We also performed KM survival analysis on the expression levels of TGF β in multiple gastric cancer datasets, and the results showed that patients with high expression of TGF β 1 in STAD

and GSE184336 had shorter survival time (Figures 2A–D), patients with high expression of TGF β 2 in STAD, GSE184336, GSE84437 and GSE62254 had shorter survival times (Figures 2E–H), and patients with high expression of TGF β 3 in GSE84437 and GSE62254 had shorter survival time (Figures 2I–L). The above results indicated that the high expression of TGF β in gastric cancer patients was an unfavorable factor for survival.

Detection of protein level of $\text{TGF}\beta$ in gastric cancer tissue

We further examined the distribution and expression of TGF β in gastric cancer tissues by IHC and Western blot assays. Firstly, immunohistochemical experiments were performed on paraffin sections of gastric cancer to observe the main distribution of TGF β 1, TGF β 2 and TGF β 3 proteins in gastric cancer tissues. TGF β 1, TGF β 2, and TGF β 3 protein expressions were distributed in similar regions, with higher levels in the cell cytoplasm, and a small amount in the tumor stroma (Figure 3A). Tissue proteins from gastric cancer and normal tissues adjacent





TGF β protein level detection. Western blot (A) and IHC (B) validation of TGF β 1, TGF β 2 and TGF β 3 proteins in gastric cancer tissues. **P < 0.01; ***P < 0.001.

to the cancer were extracted, and the results of Western blot experiments showed that the protein expression levels of TGF β 1, TGF β 2 and TGF β 3 were higher in gastric cancer tissues than in normal tissues, which was consistent with the results of transcriptome sequencing levels (GSE184336) (Figure 3B).

Relationship between different TGF β expression and clinicopathological factors in gastric cancer

TGF β expression was closely related to clinicopathological factors. TGF β 1 subgroups were associated with TNM, T and N staging in GSE184336 and with T and Histologic Grade in STAD (Supplementary Table 4). However, the clinicopathological characteristics of patients between different TGF β 2 expression subgroups were not statistically significant (Supplementary Table 5). TGF β 3 subgroups correlated with TNM, T and N stages in GSE184336 and with TNM, T and Histologic Grade in STAD (Supplementary Table 6). The combined statistical results showed that high TGF β expression was associated with poor pathological staging, such as more advanced tumor stage and poorer tumor grading.

TGF β -related gene network and TGF β signaling pathway

The GeneMANIA database was used for protein-protein interaction (PPI) network analysis of TGF β and related genes, and the KEGG database (Kyoto Encyclopedia of Genes and Genomes, https://www.kegg.jp/kegg/) demonstrated the regulatory network of the TGF β signaling pathway (TGF-beta signaling pathway - Homo sapiens (human)), and the results showed that the TGF β signaling pathway may also be involved in apoptosis and cell cycle regulation (Supplementary Figure 1).

Functional enrichment analysis in different TGF β groupings

In the GSE184336 dataset, we showed the functional expression (KEGG) matrix among different TGF β 1, TGF β 2 and TGF β 3 groupings, respectively (Figure 4A). The results showed a strong consistency in the functions involved in TGF β 1, TGF β 2 and TGF β 3 expression in different grouping situations (Figure 4B). Among the functional regulation of different TGF β factors, high expression of TGF β 1, TGF β 2 and TGF β 3 are all involved in promoting Leukocyte migration across the endothelium, ECM receptor interaction, TGF BETA signaling pathway, and VEGF signaling pathway. However, in contrast, high expression of TGF β 1, TGF β 2 and TGF β 3 all inhibited Citrate cycle (TCA cycle) (Figure 4C).

TGF β and hypoxia

Given the important role of the TCA cycle in energy supply, its physiological processes are susceptible to the influence of the oxygen environment. Therefore, we further analyzed the relationship of TGFB1, TGFB2 and TGFB3 with hypoxia in the STAD dataset. The Hallmark hypoxia score was calculated with reference to the "h.all.v7.1.entrez.gmt" gene set to evaluate the hypoxia level of patients. Correlation analysis showed that the expressions of TGF β 1, TGF β 2 and TGF β 3 were significantly positively correlated with Hallmark hypoxia (Figures 5A-C). In addition, hypoxia-related genes (ARNT, ARNT2, ARNTL, EPAS1, HIF1A, HIF3A, HK1, HK3 and PFKM) and proangiogenesis-related genes (ANGPT1, ANGPT2, FGF1, FGF2, MMP9, PDGFB, TNF and VEGFB) were highly expressed in the high TGFB1, TGFB2 and TGFB3 expression subgroups (Figures 5D-I). Therefore, hypoxia may be one of the important factors that TGF β participates in promoting gastric cancer progression.

TGF β and stromalscore, immunescore and tumor purity

Analysis of multiple gastric cancer data showed higher stromalscore and immunescore levels and lower tumor purity with high TGF β expression (Figure 6A). Therefore, in the gastric cancer microenvironment, TGF β may be involved in promoting increased mesenchymal components and immune cell infiltration.

Correlation of TGF β with immune cells

TGF β is a major regulator of multiple immune cell functions. To investigate the relationship between TGF β and immune cell infiltration in gastric cancer, we analyzed the correlation between TGF β and multiple immune cells. As shown in the figure, TGF β 1, TGF β 2 and TGF β 3 were significantly positively correlated with immune cell infiltration in multiple gastric cancer datasets. It is worth noting that due to the strong heterogeneity among gastric cancers, TGF β 1 was negatively correlated with the infiltration level of most immune cells in GSE62254. In addition, a small number of immune cells such as Activated CD4 T cells, Activated CD8 T cells and CD56dim natural killer cells were negatively correlated with TGF β 2 expression (Figure 6B). As a whole, TGF β effectively promoted immune cell infiltration in the gastric cancer microenvironment.

We further analyzed the correlation of TGF β with multiple immune cell marker genes using the TIMER database (https:// cistrome.shinyapps.io/timer/), with correlation options including none and tumor purity (Supplementary Table 7).



The results showed that TGF β 1, TGF β 2 and TGF β 3 were significantly positively correlated with most immune cell marker genes, including CD8+ T cell, B cell, Monocyte, TAM, M1 Macrophage, M2 Macrophage, Treg, T cell, Neutrophils, Dendritic cell, Th1, Th2, Tfh, Th17 and T cell exhaustion.

$TGF\beta$ and EMT

In multiple gastric cancer datasets, TGF β were positively correlated with EMT score and mesenchymal markers (CDH2, VIM and ZEB1), while significantly negatively correlated with epithelial marker CDH1 (Figure 6C). To verify the promoting effect of TGF β on EMT, we added TGF β 1, TGF β 2 and TGF β 3 active proteins to the culture medium, respectively, and cultured gastric cancer cells in TGF β environment to detect the changes of cancer cell phenotype. After 48 hours of culture, RNA was extracted from the samples. qRT-PCR results showed that CDH1 expression was decreased and CDH2, VIM and ZEB1 expression were significantly increased in the TGF β experimental group compared with the control group (Figures 7A–D), which indicated that TGF β is an important regulator of EMT and the promotion of EMT progression in gastric cancer cells.

In view of the important role of TGF β in EMT, we used small interfering RNA (siRNA) to reduce TGF β 1, TGF β 2 and TGF β 3 gene expression in gastric cancer cells. 48 hours after transfection, we extracted RNA from all samples and qRT-PCR results showed that TGF β 1, TGF β 2 and TGF β 3 expression levels were significantly reduced (Figures 8A–C). We continued the qRT-PCR analysis of the samples, and among the EMT-related markers, the expression levels of mesenchymal markers CDH2



and ZEB1 genes were significantly reduced in gastric cancer cells after TGF β gene silencing, while CDH1 expression was not significantly altered (Figures 8D–G). Thus, after silencing the TGF β gene in gastric cancer cells, the progression of EMT was significantly inhibited, and all of these results together suggest that TGF β 1, TGF β 2 and TGF β 3 are key factors in regulating the progression of EMT.

$TGF\beta$ promoted gastric cancer cell migration

We further explored the alteration of TGF β on the growth and function of gastric cancer cells. Transwell assay examined the change of gastric cancer cell migration ability in TGF β experimental group and control group (HSA), and the migration

ability of gastric cancer cells was enhanced under TGF β 1, TGF β 2 and TGF β 3 active protein stimulation conditions (Figure 7E). In contrast, the migration ability of gastric cancer cells was reduced and statistically significant after silencing of TGF β 1, TGF β 2 and TGF β 3 genes (Figures 8H–J). Therefore, TGF β has a strong promoting effect on cancer cell value addition and migration. Considering the significant suppression of EMT trend after TGF β gene silencing, EMT may play an indispensable role in the regulation of gastric cancer cell migration by TGF β .

TGF β with TMB and MATH

We first showed the mutation information of TGF β and related genes, among which ACVR2A, CREBBP, SMAD4, BMPR2 and CTCF had higher mutation frequencies. Mutation



exclusive and co-occurrence among mutant genes were mainly concentrated in genes with high mutation frequency (ACVR2A, CREBBP, SMAD4, BMPR2, CTCF, SPTBN1, EP300 and BMPR1A) (Figures 9A, B). In addition, TGF β 1 and TGF β 3 expression were negatively correlated with Mutant-allele tumor heterogeneity (MATH) (Figures 9C–E). We also compared the tumor mutation burden (TMB) between different TGF β groups and showed that TMB was higher in the TGF β 1, TGF β 2 and TGF β 3 low expression groups (P < 0.05), while higher TMB was beneficial for prolonging patient survival (Figures 9F–I). Therefore, the high expression of TGF β in the gastric cancer microenvironment may be an important factor in promoting tumor differentiation and poor prognosis.

Analysis of copy number variation between different TGF β groups and immune cell infiltration

We showed the differences in copy number variants (CNV) in different TGF β groupings (Supplementary Figures 2A, C, E) and also analyzed the effect of somatic copy number alterations (SCNAs) of TGF β on immune cell infiltration to elucidate the potential mechanism of TGF β associated with immune cell infiltration. Arm-level deletion in TGF β 1-associated SCNAs was significantly associated with the level of B-cell, CD4+ T-cell, CD8+ T-cell, neutrophil, macrophage, and dendritic cell infiltration. In contrast, arm-level gain in TGF β 2- and TGF β 3-related SCNAs had a greater effect on B cells, CD4+ T cells, CD8+ T cells, neutrophils, macrophages and dendritic cell infiltration (Supplementary Figures 2B, D, F).

TGF β and immunotherapy

Since TGF β is involved in several aspects of the tumor development process, we further analyzed the relationship between TGF β and immunotherapy for gastric cancer. TGF β was significantly positively correlated with immune checkpoint and MHC molecules in the STAD dataset (Supplementary Figure 3A), and PD-1, PD-L1 and CTLA-4 expression levels were higher in the high TGF β subgroup (Figures 10A–C, E–G, I–K). We used two methods to assess immunotherapy response. First, the submap module results showed a response to CTLA-4 immunotherapy in the high TGF β 1, TGF β 2 and TGF β 3 subgroups, and a response to PD-1 immunotherapy in the high TGF β 1 subgroup (Figures 10D,



H, L). In addition, using the ImmuCellAI database to further assess the guiding significance of TGFβ for ICB therapy, with high TGFβ1 (χ 2 = 17.739, *P* < 0.001), TGFβ2 (χ 2 = 7.978, *P* = 0.005) and TGFβ3 (χ 2 = 8.237, *P* = 0.004) subgroups had a higher proportion of patients responding to immunotherapy (Figures 10M–O). However, within multiple gastric cancer datasets including GSE184336, the correlation of TGFβ with immune checkpoint and MHC molecules differed (Supplementary Figure 3A), but TGFβ still had a guiding effect on immunotherapy response (Supplementary Figures 3B–D). Therefore, TGFβ still needs to be analyzed and validated in more gastric cancer data sets before it can be used as a biomarker for predicting responsiveness.

TGF β guidance for chemotherapy

As an important method of adjuvant treatment for gastric cancer, chemotherapy drug therapy occupies an important position in clinical treatment. Prediction of differences in chemical drug IC50 between different TGF β subgroups according to the GDSC database showed increased IC50 for most chemicals in the high TGF β 1, TGF β 2 and TGF β 3 expression subgroups of the STAD and GSE184336 datasets (Figure 11; Supplementary Figure 4).

Discussion

In this study, we analyzed the transcriptomic dataset GSE184336 of 231 gastric cancer patients, and the results showed that TGF\$1, TGF\$2 and TGF\$3 were highly expressed in cancer tissues, and western blot assays further confirmed the differences in TGFB expression. Pathological factor analysis between different TGF β groupings in STAD and GSE184336 showed that $\mbox{TGF}\beta$ was associated with poorer pathological staging or grading, suggesting that $TGF\beta$ may be an important factor in the poor progression of gastric cancer. Survival analysis of GSE184336 showed shorter survival times in patients with high TGFB1 and TGFB2 expression, and similar results were seen in several gastric cancer datasets set. Colorectal cancer patients have high TGFB1 levels compared to healthy controls, and high levels of TGFβ1 are positively correlated with advanced tumor stage and metastasis after surgical resection (33, 34). In addition, high levels of TGFB2 were a factor of poor prognosis in patients with gastric cancer, which is consistent with the results reported in previous studies (35, 36). TGFB3 was also a poor prognostic factor for STAD in this study, however, compared to the large amount of data available for TGF β 1, there is a lack of relevant data demonstrating the pathogenic role of TGFB3 in tumorigenesis, so this study is an important addition to the role of TGFβ3 molecules in gastric cancer.



Reviewing the role and expression of TGF β in different types of cancer, we found that TGF β is highly expressed in most types of cancer and is one of the risk factors affecting cancer prognosis, however the mechanism by which TGF β affects gastric cancer progression remains to be clearly defined (37–39). In the GSE184336 dataset, GSVA analysis of patients with different TGF β 1, TGF β 2 and TGF β 3 subgroups showed that high TGF β expression significantly inhibited the TCA cycle, which plays a key role in energy metabolism and is closely related to the tissue oxygen environment. Therefore, we further analyzed the relationship between TGF β 1, TGF β 2 and TGF β 3 were significantly and positively correlated with Hallmarks hypoxia, reflecting a hypoxic microenvironment with high TGF β factors, meanwhile the hypoxia-related genes HIF gene family and proangiogenic genes were highly expressed in the TGF β high expression group. Hypoxia-inducible factor (HIF) is the main transcriptional regulator in response to hypoxia and consists of HIF- α subunits (HIF-1 α or HIF-2 α) and HIF-1 β under hypoxic conditions (40). In earlier studies, HIF-1 α was associated with TGF β activation in hepatocytes and human umbilical vein endothelial cells during hepatic fibrosis, and TGF β also inhibited mRNA and protein expression of PHD2, thereby increasing the stability of HIF-1 α (41, 42). Overexpression of HIF-1 α in breast cancer promotes the expression of TGF β 1 and SMAD3 (43). Endothelial cells in hypoxia [1% partial pressure of oxygen (PaO2)] increase messenger RNA and protein levels of TGF β 2, as well as messenger RNA levels of type II membrane



receptors for TGF-beta2 (44). Therefore, the involvement of TGF β signaling pathway in the regulation of hypoxia may be one of the important factors promoting tumor progression.

EMT is the process by which polar epithelial cells convert to migratory mesenchymal cells and gain the ability to invade and migrate, and it is present in several physiological and pathological processes in the human body (45, 46). As cancer cells diminish their epithelial characteristics during EMT, they may express fewer tumor-specific neoantigens to avoid recognition by immune cells, all of which contribute to cancer progression, with TGF β being a key factor in EMT regulation (47, 48). The results of this study showed that TGF β 1, TGF β 2 and TGF β 3 were positively correlated with EMT, CDH2, VIM and ZEB1 and significantly negatively correlated with CDH1 in multiple gastric cancer datasets. To further verify the regulatory role of TGF β in gastric cancer cells, we cultured gastric cancer cells with conditioned medium incorporating TGF β 1, TGF β 2 and TGF β 3 active proteins and showed that the EMT trend of cancer cells cultured under TGF β conditions was significantly increased (CDH1 expression was decreased, CDH2, VIM and ZEB1 expression was increased), while inhibition of gastric cancer cells TGF β 1, TGF β 2 and TGF β 3 gene expression suppressed the EMT trend in gastric cancer cells (no significant change in CDH1 expression and decreased expression of CDH2 and ZEB1). It further demonstrated the important role of TGF β gene family on the regulation of EMT.

In the gastric cancer microenvironment, TGF β was significantly positively correlated with stromalscore and immunescore, and negatively correlated with tumor purity, and similar results were well reflected in multiple gastric cancer data sets. Another



important aspect in TME was to explore the relationship between TGF β and immune infiltration level of gastric cancer. The results showed that in gastric cancer TGF β 1, TGF β 2 and TGF β 3 expressions were strongly correlated with most of the immune cell infiltration levels. The relationship between TGF β and immune cell infiltration was also shown for gene mutations, and somatic copy number alterations (arm-level deletion and arm-level gain) of TGF β gene had a greater impact on immune cell infiltration. In the present study TGF β was significantly and positively correlated with Tregs, which promote the formation of an immunosuppressive microenvironment and attenuate the antitumor effects produced by CD4+ T cells, CD8+ T cells and NK cells through secreted TGF β (49, 50). It has been demonstrated that a decrease in the number of Tregs in different mouse models significantly increased the

antitumor immune effect in mice (51). Dysfunction of anti-tumor immune cells in tumor patients is also closely related to Tregs, with a significant increase in the number of Tregs within and at the margins of tumor tissues such as gastric cancer, breast cancer, and melanoma (52–54). In TME, although the level of infiltration of immune cells with anti-tumor capacity increases with increasing levels of TGF β expression, it is usually accompanied by a compensatory increase in immunosuppressive cells.

The results of this study showed that patients in the low TGF β expression group had higher TMB, and patients with higher TMB had longer survival. Recent studies have shown that high TMB increases the likelihood that immunogenic neoantigens expressed by tumor cells induce a response to immunotherapy. In addition, TGF β 1 and TGF β 3 were



TGFβ3 groups.

negatively associated with MATH, which is prevalent in most cancer patients and is a major driver of acquired resistance to cancer therapy (55, 56). In the context of immunotherapy, the pressure on the immune system to respond to specific tumor antigens can drive selection against antigen-negative cells, which is a common cause of clinical relapse. In the present study MATH was lower when TGF β 1 and TGF β 3 were higher, while higher levels of TGF β promoted the formation of an immunosuppressive microenvironment and facilitated the progression of EMT, conditions that favored the evolutionary development of cancer cells. Stromal fibroblasts and other cells in tumor tissues shape the immunosuppressive environment of tumors through TGF β signaling, inhibiting the antitumor activity of immune cells and preventing or weakening the effect of anticancer immunotherapy (57). Therefore, inhibition of TGF β signaling is considered as a prerequisite and an important way to improve the effectiveness of immunotherapy. Considering TGF β , CTLA4 and PD-L1/PD-1 as parallel immunosuppressive pathways, combining TGF β inhibitors with other immune checkpoint inhibitors may improve the therapeutic efficacy (58–61). The results of this study showed that the expression levels of immune

checkpoints PD-1, PD-L1 and CTLA-4 were higher in the high TGF β subgroup, and in addition we predicted the response of TGF β expression levels to ICB therapy, with a higher proportion of patients in the high TGF β subgroup responding to ICB therapy. Combination therapy has been pre-evaluated in mouse cancer models where, according to the model and experimental design, therapeutic co-administration of TGF β blockade and anti-PD-L1 antibodies reduced TGF β signaling in stromal cells, promoted T-cell infiltration into tumor centers, and provoked potent antitumor immunity and tumor regression (62).

Furthermore, in drug prediction of patients with different TGF β subgroups, it was found that patients in the high TGF β group were less sensitive to treatment with small molecule compounds, which may be related to the increased extracellular interstitial component of gastric cancer induced by high TGF β levels. The rigid Extracellular matrix (ECM) that forms around the tumor reduces the spread of therapeutic agents to cancer cells, while the dense ECM reduces the vascular density and causes the vessels to embed in the matrix, forming a tough barrier that cannot be perfused with drugs (63).

However there are many clinical challenges in developing TGF β inhibitors, especially patient selection and timing of treatment. Considering the dual role of TGF β on proliferation, TGF β inhibitors may be beneficial in advanced tumors. It is worth noting that TGF β plays a positive factor in SKCM, KIRC and THYM in pan-cancer analysis and should be considered carefully when using TGF β -related signaling inhibitors. Before TGF β inhibitors can be used clinically, a lot of research on different types of cancer is still needed. To comprehensively evaluate the efficacy of TGF β inhibitors in immunotherapy in different cancer types and cancer stages, and whether they need to be used in combination with other immune target inhibitors. Only then can we make accurate screening and evaluation of the target population and therapeutic efficacy of TGF β inhibitors.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found in the article/ Supplementary Material.

Ethics statement

The studies involving human participants were reviewed and approved by Research Ethics Committee of the Affiliated Cancer Hospital of Harbin Medical University. The patients/ participants provided their written informed consent to participate in this study.

Author contributions

BH and TF conceived the project and wrote the manuscript. BH and YZ participated in data analysis. JG and YLZ participated in discussion and language editing. YX reviewed the manuscript. All authors contributed to the article and approved the submitted version.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fonc. 2022.920599/full#supplementary-material

SUPPLEMENTARY FIGURE 1

TGF β network analysis. (A) Using GeneMANIA to study functional associations between different proteins, the colors of the connections represent different correlations, and the different colors in the circles represent the different functions involved. (B) Regulation of TGF β signaling pathway.

SUPPLEMENTARY FIGURE 2

Analysis of copy number variation among different TGF β groups. (A, C, E) Comparison of the frequency of copy number changes in different TGF β groupings. Chromosomal locations of peaks of significantly recurring focal amplification (red) and deletions (blue) were presented. (B–D) Comparison of tumor infiltration levels between tumors with different somatic copy number alterations of TGF β . * P < 0.05; ** P < 0.01; ***P < 0.001.

SUPPLEMENTARY FIGURE 3

TGF β with MHC molecules and Immune checkpoints. (A) Correlation of TGF β with MHC molecules and Immune checkpoints. (B–D) The expression levels of TGF β 1, TGF β 2 and TGF β 3 in different immunotherapy response subgroups. * P < 0.05; ** P < 0.01; **** P < 0.001.

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SUPPLEMENTARY FIGURE 4

TGFβ and chemotherapy in GSE184336. (A) Boxplots depicted the differences in the estimated IC50 levels of AKT.inhibitor.VIII, BIBW2992, BMS.708163, Gefitinib, GW.441756, Lapatinib, Metformin, PF.4708671, Roscovitine and Sorafenib between the high and low TGFβ1 groups. (B) Boxplots depicted the differences in the estimated IC50 levels of AKT.inhibitor.VIII, AZD6244, BIBW2992, Erlotinib, Gefitinib, GW.441756, Lapatinib, Metformin, Roscovitine and Sorafenib between the high and low TGFβ2 groups. (C) Boxplots depicted the differences in the estimated IC50 levels of A.443654, AKT.inhibitor.VIII, BIBW2992, Gefitinib, Lapatinib, Metformin, Paclitaxel, PF.4708671, Roscovitine and Sorafenib between the high and low TGFβ3 groups.

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